

# computational considerations and apparently awesome alliterations

-

16. joulukuuta 2020

# what about that quantum

"Quantum mechanics is a trivial theory" -Unnamed Instructor

$$\hat{H}_{JC} = \frac{\hbar\Omega}{2}(\hat{a}\hat{\sigma}_+ + \hat{a}^\dagger\hat{\sigma}_-) + \hbar\omega_c\hat{a}^\dagger\hat{a} + \hbar\omega_a\frac{\hat{\sigma}_z}{2} \quad (1)$$

$$\hat{\sigma}_+ = |g\rangle\langle e| \quad (2)$$

$$\hat{\sigma}_- = |e\rangle\langle g| \quad (3)$$

stuff in hilbert space = matrices

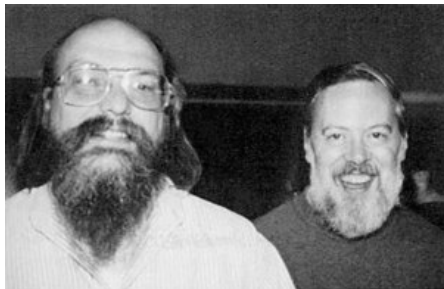
tensor product = matrices

everything is matrices

# matrices: a solved problem

## nice things about matrices

- ▶ There are numerous libraries available for matrix operations (didn't use them tho)
- ▶ Written by wizards in the 70s (see picture)
- ▶ Run faster than me confronted with the prospect of having to socialize



Kuva: Average numerical linear algebra alpha male

# matrices: a solved problem

infinite matrices??

$$\hat{a} = \frac{1}{\sqrt{2}}(\hat{x} + i\hat{p}) \quad (4)$$

$$\hat{a}^\dagger = \frac{1}{\sqrt{2}}(\hat{x} - i\hat{p}) \quad (5)$$

matrix rep :

$$\hat{a} = \begin{bmatrix} 0 & 1 & 0 & 0 & \dots \\ 0 & 0 & \sqrt{2} & 0 & \dots \\ 0 & 0 & 0 & \sqrt{3} & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix} \quad (6)$$

# matrices: a solved problem

truncation and error

$$\hat{a} = \begin{bmatrix} 0 & 1 & 0 & 0 & \dots \\ 0 & 0 & \sqrt{2} & 0 & \dots \\ 0 & 0 & 0 & \sqrt{3} & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix}, \hat{a}^\dagger = \begin{bmatrix} 0 & 0 & 0 & 0 & \dots \\ 1 & 0 & 0 & 0 & \dots \\ 0 & \sqrt{2} & 0 & 0 & \dots \\ \vdots & \vdots & \ddots & \vdots & \ddots \end{bmatrix}$$

need to impose a cutoff at some  $n$ , but how does error decrease?

ERROR = % OF ELEMENTS THAT DON'T MATCH

# matrices: a solved problem

truncation and error

$$[\hat{a}, \hat{a}^\dagger] = 1 \quad (7)$$

Matrix rep,  $n = 5$  and  $n = 6$

$$[\hat{a}, \hat{a}^\dagger] = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}, [\hat{a}, \hat{a}^\dagger] = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

error = 20% and = 18%

# matrices: a solved problem

## truncation and error

so: error decreases as  $O(n)$  but matrix size increases as  $O(n^2)$ .

```
for(int i = 0; i < rows; i++) {  
    for(int j = 0; j < columns, j++) {  
        ...  
    }
```

so execution time:  $O(n^2)$ , sometimes worse. not ideal, can't be helped!

# Why are QM problems computationally difficult?

## Computer memory

computer is a stupid rock tricked in to thinking!

memory is aligned, meaning:

- ▶ processor starts reading RAM at addresses divisible by 64(8)
- ▶ 4-byte variables must be in a memory slot divisible by 4, 6 byte variables in spots divisible by 6 etc
- ▶ reading RAM is S L O W



# Why are QM problems computationally difficult?

## Computer memory

illustration:

```
struct Person {  
    char first_letter_of_name; // 1 byte  
    int age;                   // 4 bytes  
    short number_of_kids;      // 2 bytes  
}
```

```
struct Person {  
    char first_letter_of_name; // 1 byte  
    char padding[3]            // 3 bytes  
    int age;                   // 4 bytes  
    short number_of_kids;      // 2 bytes  
}
```

in code: 7 bytes = single read by CPU

in memory: 10 bytes = two reads by CPU = 2x slower!

it takes 2-10 ns to access CPU cache, 200+ns to access RAM

fix: "memory packing" -> reorder the elements according to alignment. NOT POSSIBLE in anything except low level languages  
QM = huge amounts of data, thus huge amounts of accesses, thus a difficult computational problem

# Example 1

## Semiclassical Rabi

$$\dot{C}_l(t) = -\frac{i}{\hbar} \sum_k C_k(t) \langle l | H' | k \rangle \quad (8)$$

in the code (under the rotating wave approximation):

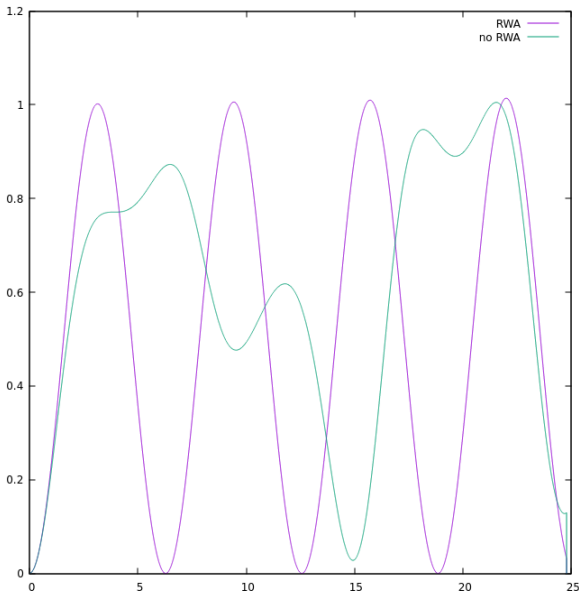
```
complex double C1 = l/2 *matmul(p,b0,matmul(p,d,k1))  
->matrix[0][0]*x2*cexp(l*detuning*t);  
complex double C2 = l/2 *matmul(p,b1,matmul(p,d,k0))  
->matrix[0][0]*x*cexp(-l*detuning*t);
```

also, python code to interactively run the simulation:

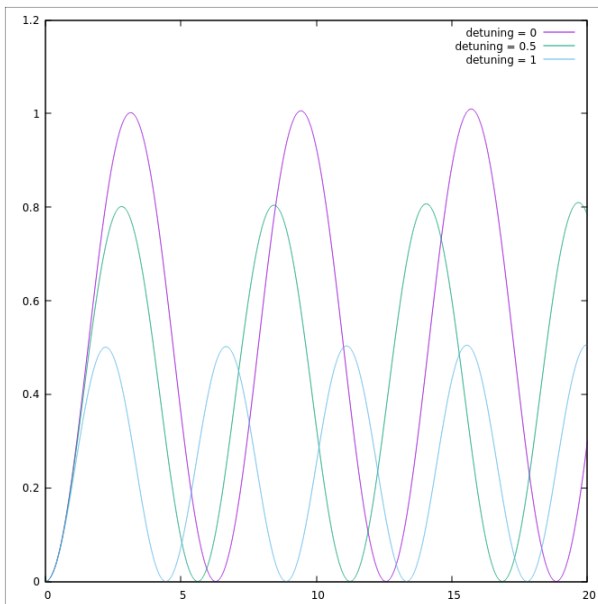
```
import scqo
system = scqo.NLevelAtom(omega = 0.5)
field = scqo.ClassicalField(omega = 0.5)
system.calculate()
#no rwa
system.set_rwa(False)
system.calculate()
```

Plot the results:

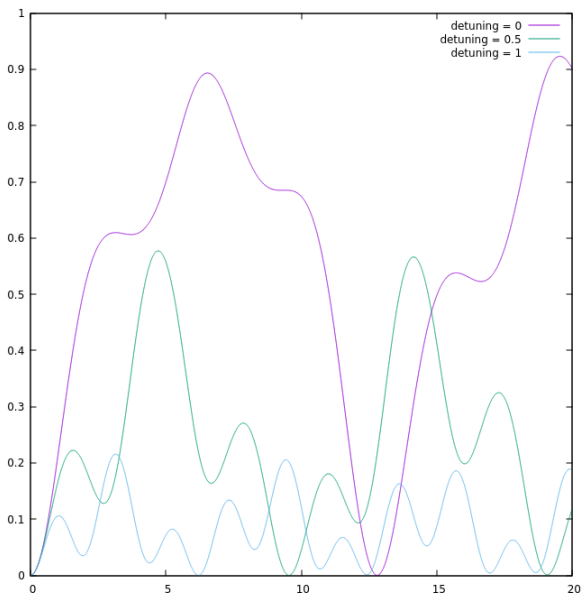
lesson about approximations: they might actually suck. cows are not always spherical



## Different detunings.



..with no RWA



## Example 2

Bose-Hubbard dimer (=two sites)

Bose-Hubbard timer: lattice sites + particles interacting.

$$\hat{H} = \epsilon(\hat{a}_1^\dagger \hat{a}_1 - \hat{a}_2^\dagger \hat{a}_2) + v(\hat{a}_1^\dagger \hat{a}_2 + \hat{a}_2^\dagger \hat{a}_1) + c(\hat{a}_1^\dagger \hat{a}_1 - \hat{a}_2^\dagger \hat{a}_2)^2 \quad (9)$$

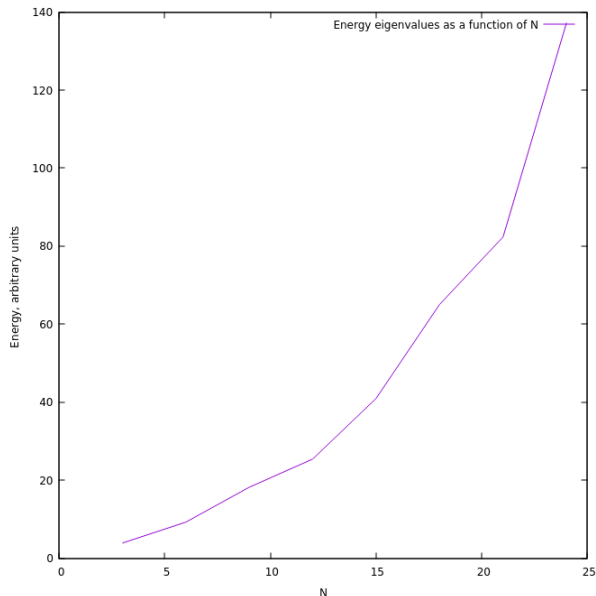
huge matrix size for larger particle numbers, takes forever to run.

easy to implement with matrix methods



# Example 2

Bose-Hubbard dimer (=two sites)



computational note: you can remove 0-padding from Hamiltonians without losing any important information about the energy levels:

$$H = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & a & b & c & d & 0 \\ 0 & f & b & c & d & 0 \\ 0 & a & p & q & d & 0 \\ 0 & a & b & c & d & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \iff H = \begin{bmatrix} a & b & c & d \\ f & b & c & d \\ a & p & q & d \\ a & b & c & d \end{bmatrix} \quad (10)$$

Only zeroed eigenvalues are lost, and there are  $2n$  of them, where  $n$  is the amount of padding removed.

## Example 3: Simulating dynamics

### Bose-Hubbard Lindblad simulation

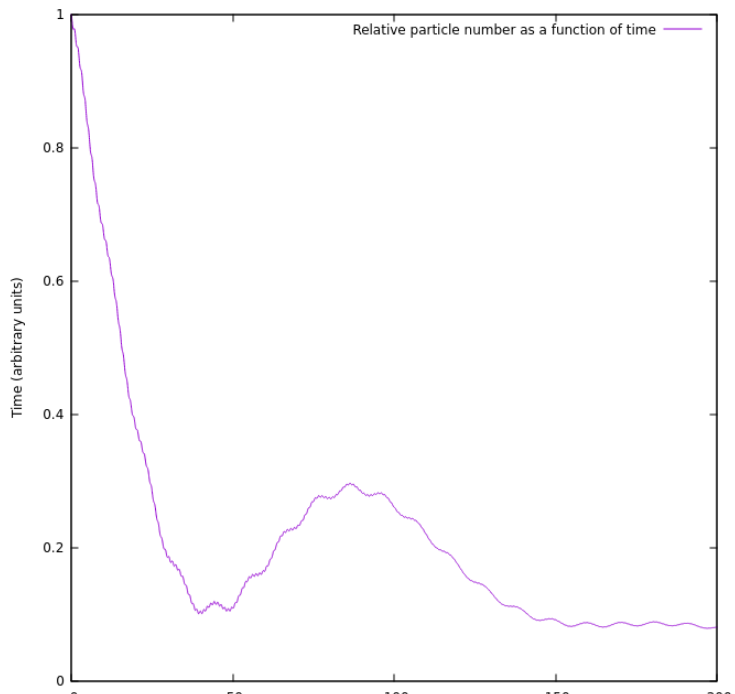
The Bose-Hubbard model as an open system (Lindblad form):

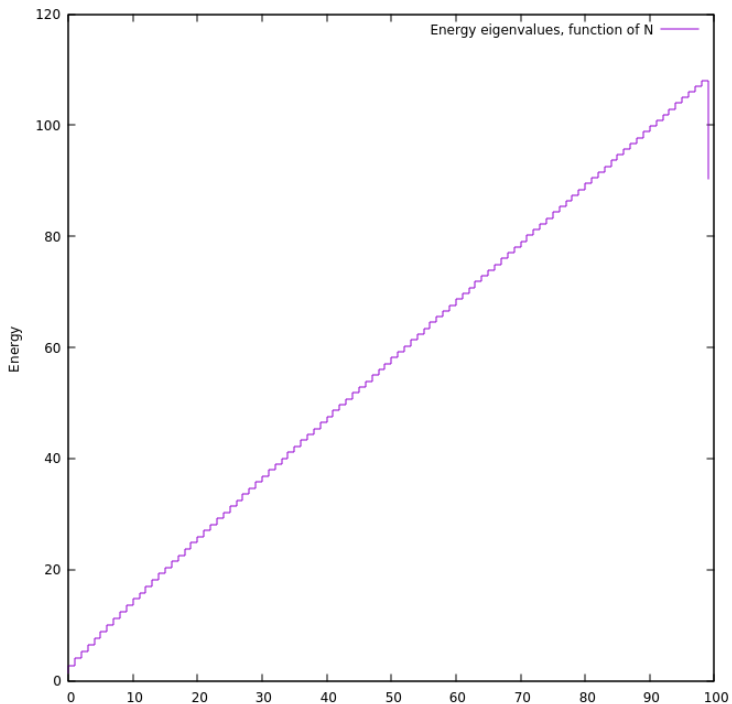
$$\dot{\hat{\rho}} = -i[\hat{H}, \hat{\rho}] - \frac{\gamma}{2} \left( \hat{a}_2^\dagger \hat{a}_2 \hat{\rho} + \hat{\rho} \hat{a}_2^\dagger \hat{a}_2 - 2\hat{a}_2 \hat{\rho} \hat{a}_2^\dagger \right) \quad (11)$$

$$\hat{H} = \epsilon(\hat{a}_1^\dagger \hat{a}_1 - \hat{a}_2^\dagger \hat{a}_2) + v(\hat{a}_1^\dagger \hat{a}_2 + \hat{a}_2^\dagger \hat{a}_1) + c(\hat{a}_1^\dagger \hat{a}_1 - \hat{a}_2^\dagger \hat{a}_2)^2 \quad (12)$$

H is the Hamiltonian discussed previously. How about dem dynamics?

(optical tunneling from the dimer to optical lattice)





## Some conclusions

- ▶ Matrices are a straightforward way of doing OQS
- ▶ The method is a bit slow (could be heavily optimized)
- ▶ Special care is needed to choose the right algorithms for unstable problems
- ▶ Approximations sometimes suck
- ▶ I hate segfaults